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NEW RESULTS IN 2-D SYSTEMS THEORY, 2-D STATE-SPACE MODELS - REALIZATION
AND THE NOTIONS OF CONTROLLABILITY, OBSERVABILITY AND MINIMALITY

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Abstract

A short comparison between the different state-space models is presented. We discuss proper definitions of state, controllability and observability and their relation to minimality of 2-D systems. We also present new circuit realizations and 2-D digital filter hardware implementations of 2-D transfer functions, as well as a 2-D generalization of Levinson's algorithm.

1. Introduction

Attasi, Fornasini-Marchesini, Givone-Roeser have proposed different state-space models for 2-D systems and have suggested some extensions of the usual 1-D notions of controllability, observability and minimality to the 2-D case. However, these results are not quite satisfactory; they either lack motivation for the state-space models introduced or the notion of state-space is improperly defined.

In this paper and in [18], we tried to provide answers to these questions from a practical as well as algebraic standpoint. We start with a discussion of all the current models based on a practical (circuit-oriented) point of view and on a proper definition of state. Since the other models can be imbedded in the Givone-Roeser model, it appears to be the most satisfactory one.

From the circuit point of view, we present in Section 3 an implementation of 2-D transfer functions using two types of dynamic elements - horizontal delay elements z^{-1} and vertical delay elements w^{-1} . The hardware implementation of 2-D digital filters for imaging systems is also discussed in Section 3.

An algebraic approach based on eigencurves and eigencones enables us in Section 5 to introduce the concept of modal controllability (observability). We show that a system is minimal if and only if it is modally observable and controllable.

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If we are given an irreducible transfer function of order (n,m) , a state-space realization is minimal if and only if it is of size $n+m$. The existence of such $(n+m)$ (real or complex) state realizations is discussed in Section 6, where we provide a simple counter example to a real model. In the Appendix we give a 2-D generalization of Levinson's algorithm. In conclusion, it appears that the results obtained by the algebraic and the practical approaches are quite compatible.

2. State-Space Models for 2-D Systems

During recent years, several authors: Attasi [1], [2], Fornasini-Marchesini [3], [4] and Givone-Roeser [5] have proposed different state-space models for 2-D systems. In [4], Fornasini and Marchesini were using the algebraic point of view of Nerode equivalence and were the first to realize that a major difference between 1-D and 2-D systems is that we can introduce a global state and a local state in the 2-D case. The global state (which is of infinite dimension in general) preserves all the past information while the local state gives us the size of the recursions to be performed at each step by the 2-D filter. However, their state does not obey a first-order difference equation (the notion of first order difference equation for linear systems on partially ordered sets has been defined by Mullans and Elliott in [7]). Attasi's model [1], [2] suffers from the same drawback.

Givone and Roeser in [8] and [5] have used a "circuit approach" to the problem of state space realization for some 2-D transfer functions. They present a model in which the local state is divided into an horizontal and a vertical state which are propagated respectively horizontally and vertically by first order difference equations.

Mitra, Sagar and Pendergrass gave a realization for arbitrary transfer functions in [9] by presenting an implementation method for 2-D transfer functions using some delay elements z^{-1} and w^{-1} via an approach that is consistent with Roeser's model. A detailed comparison can be found in [16], Part II.

3. Circuit Realizations and Hardware Designs

First, we can note that the notion of "dynamic elements", "multipliers" and "adders" is at the center of circuit theory. In the 1-D discrete-time case, the dynamic elements used are (time) delay

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elements. The 1-D realization problems have been well studied and, given any transfer function, it is well known that the realization can be readily found in certain standard (e.g. controller canonical) forms [11]. For the realization of a 2-D transfer function, a major difference is that two types of dynamic elements are needed - "horizontal delay element" (z^{-1}) and "vertical delay element" (ω^{-1}). Now an important problem is that of how to use 2-D dynamic elements, multipliers and adders to realize a 2-D digital filter with the transfer function:

$$H(z^{-1}, \omega^{-1}) = \frac{b(z^{-1}, \omega^{-1})}{a(z^{-1}, \omega^{-1})} = \frac{\sum_{i=0}^n \sum_{j=0}^m b_{ij} z^{-i} \omega^{-j}}{\sum_{i=0}^n \sum_{j=0}^m a_{ij} z^{-i} \omega^{-j}} \quad (3.1)$$

We can do this in two steps. First we rewrite (3.1) in a rational-gain representation, i.e.

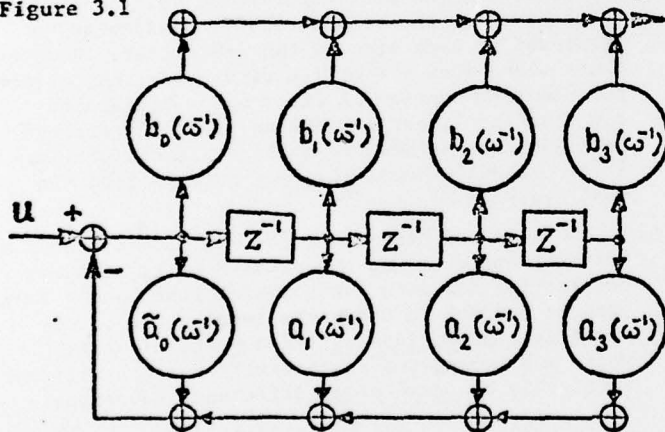
$$H(z^{-1}, \omega^{-1}) = \frac{\sum_{i=0}^n b_i(\omega^{-1}) z^{-i}}{\sum_{i=0}^n a_i(\omega^{-1}) z^{-i}} \quad (3.2)$$

Without loss of generality, we can assume $a_{00} = 1$ and we denote

$$a_0(\omega^{-1}) \triangleq 1 + \tilde{a}_0(\omega^{-1})$$

Thus, using the 1-D realization technique, we write down a realization, where the gains of the multipliers are represented in $F[\omega^{-1}]$.

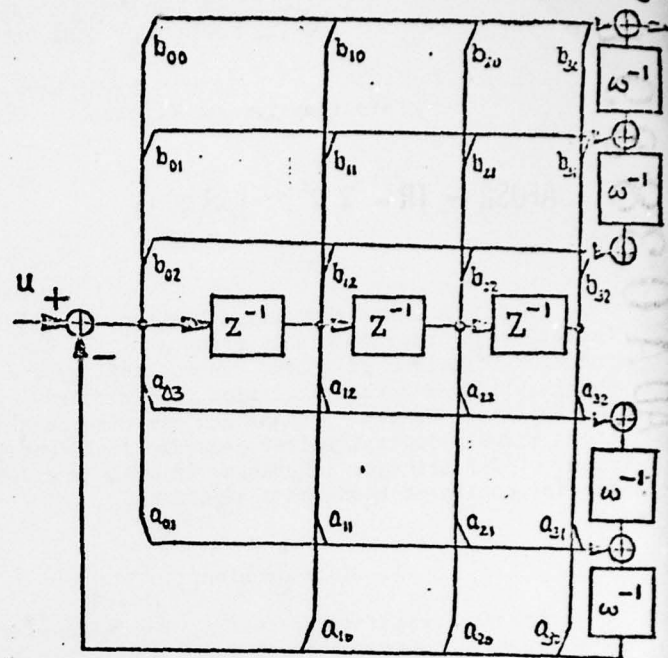
Figure 3.1



The realization is almost achieved: in addition to the n horizontal delay elements we need only m vertical delay elements to implement the feedback gains $\{a_i(\omega^{-1}), i=0, 1, \dots, m\}$ and m other vertical delay elements to implement the readout gains $\{b_i(\omega^{-1}), i=0, 1, \dots, m\}$. Thus, the complete realization shown in Figure 3.2 requires only $n+2m$ dynamic elements.

This realization is a standard (canonical) one; its structure is very simple and it involves only real gains. Note also that we need fewer dynamic elements than the implementations in [9].

Figure 3.2: 2-D Controller Form Realization



State Space Model Representation

As remarked in Section 2, circuit implementations with delay elements z^{-1} and ω^{-1} are in a one-to-one correspondence with state space models of Roeser's type. The output of the z^{-1} delays are the horizontal states and the outputs of the ω^{-1} delays are the vertical states.

Thus, the implementation of Figure 3.2 can be transformed readily into the following state space model:

$$\begin{bmatrix} x_h(i+1, j) \\ x_{v_1}(i, j+1) \\ x_{v_2}(i, j+1) \end{bmatrix} = \begin{bmatrix} x_h(i, j) \\ x_{v_1}(i, j) \\ x_{v_2}(i, j) \end{bmatrix} + b u(i, j) \quad (3.3)$$

$$y(i, j) = c x(i, j) + b_{00} u(i, j)$$

where

$$A = \begin{bmatrix} A_{11} & \dots & -e_1 e_1' & \dots & 0 \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ \tilde{A} & \dots & A_{22} & \dots & 0 \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ B & \dots & -b_{0m} e_1' & \dots & Z \end{bmatrix}, \quad \begin{aligned} b' &\triangleq [e_1', a_{0m}', b_{0m}'] \\ c &\triangleq [b_{n0}', -b_{00} e_1', e_1'] \end{aligned}$$

with

$$A_{11} \triangleq Z_n' - e_1 a_{n0}', \quad A_{22} \triangleq Z_m' - a_{0m} e_1'$$

and

$$\begin{aligned} [\tilde{A}]_{ij} &\triangleq a_{ij} - a_{i0} a_{0j}, \quad (1, 1) \leq i, j \leq (n, m); e_1 \triangleq [1, 0, \dots, 0]' \\ [\tilde{B}]_{ij} &\triangleq b_{ij} - a_{i0} b_{0j}, \quad 1 \leq i \leq n, \quad 0 \leq j \leq m; b_{0m} \triangleq [b_{01}, \dots, b_{0m}]' \\ [Z]_{ij} &\triangleq \begin{cases} 1 & \text{if } i=j+1 \\ 0 & \text{else} \end{cases}; \quad \begin{aligned} a_{n0} &\triangleq [a_{n1}, \dots, a_{nm}]' \\ a_{0m} &\triangleq [a_{01}, \dots, a_{0m}]' \end{aligned} \end{aligned}$$

Hardware Design of 2-D Digital Filter

The idea of using two types of dynamic elements is not very abstract; it is very natural in delay-differential systems. However, before considering its practical application to image systems, two remarks have to be made:

1) Because the "spatial" dynamic elements seem unimplementable, (except as index operators in a digital computer, for example,) we can replace them by time-delay elements.

2) In order to have a finite order description, we shall only consider a bounded frame system, i.e. we assume that the picture frame of interest is an $M \times N$ frame (with vertical width M and horizontal length N).

Note that in order to use time delay elements we need first to find a way to code a 2-D spatial system into a 1-D (discrete-time) system and vice versa. Thus we shall propose the following implementation of a 2-D filter:

i) The input scan generator codes the 2-D spatial input into 1-D (time) data according to the mapping function $t(\cdot, \cdot)$

$$t(i, j) = iM + jN \quad (3.4)$$

where M and N are relatively prime integers.

ii) A 1-D (discrete-time) digital filter processes the 1-D data generated by (i). This subsystem is implemented by replacing z^{-1} by δ , ω^{-1} by Δ in a 2-D circuit realization (e.g. 2-D controller form). δ and Δ are chosen as

$$\delta = D^M = M\text{-units delay element}, \quad (3.5)$$

$$\Delta = D^N = N\text{-units delay element}.$$

iii) The output frame generator decodes the 1-D (discrete-time) output of the 1-D digital filter described above into a 2-D (discrete-spatial) picture according to the inverse mapping of (3.4).

$$(i(t), j(t)) = (Pt \bmod M, [t - (Pt \bmod M)N]/M) \quad (3.6)$$

where P is the unique integer such that

$$PN - QM = 1 \text{ and } 0 < P < M \quad (3.7)$$

Verification: Let us note the 1-D (discrete-time) output will be

$$\begin{aligned} y(D) &= H(D) U(D) \\ &= H(z^{-1}, \omega^{-1}) u(z^{-1}, \omega^{-1}) \Big|_{z^{-1}=D^M, \omega^{-1}=D^N} \\ &= \sum_i \sum_j y_{i,j} z^{-i} \omega^{-j} \Big|_{z^{-1}=D^M, \omega^{-1}=D^N} \end{aligned} \quad (3.8)$$

where $\{y_{i,j}\}$ represents the 2-D (discrete-spatial) output data field. Note also that

$$y(D) \triangleq \sum_t y_t D^{-t} \quad (3.9)$$

$$y_t = \sum_{(i,j): iM+jN=t} y_{i,j} \quad (3.10)$$

Since the system is a causal system,

$$y_{i,j} = 0 \quad \text{if} \quad i, j < 0 \quad (3.11)$$

Let us consider only the integer t with

$$t = iM + jN, \quad 1 \leq N, j \leq M$$

then (3.10) and (3.11) give

$$y_t = y_{i,j}$$

since, for this special case, the summation set of (3.10) contains only one nonzero point. Therefore, we will obtain a bona fide output picture inside the $M \times N$ frame.

This 2-D image scanning and display system is not as complicated as it looks, it can be simple:

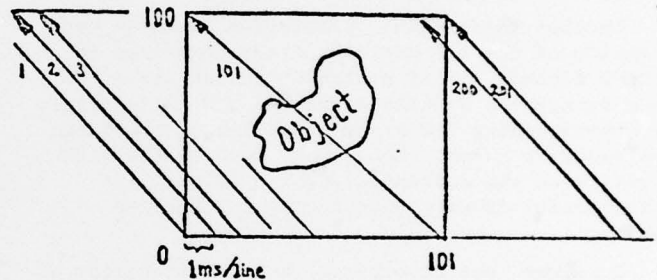
Example: Problem: Design a 2-D digital filter for

$$H(z^{-1}, \omega^{-1}) = \frac{1}{1 + 0.3z^{-1} + 0.2\omega^{-1} + 0.1z^{-1}\omega^{-1}}$$

for a frame: $M \times N = 100 \times 101$. Assume $D = 0.01$ ms.

Solution. (i) ISG

In this special frame (with $N=M+1$), the input scanning generator is indeed very simple, as shown in Figure 3.3.

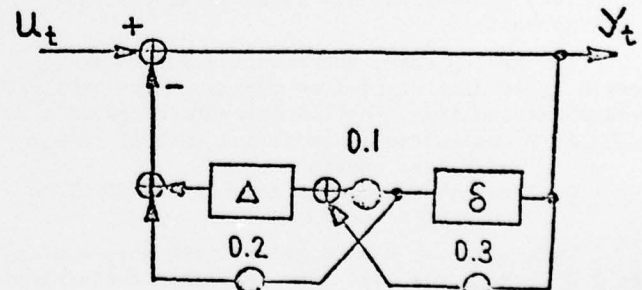


Scanning time: 0.01 ms/pixel = 1 ms/line
Scanning angle: 45°

Fig. 3.4 Input Scan Generator & Output Frame Generator

(ii) 1-D digital filter

Constructing the 2-D realization of Figure 3.2 and then replacing z^{-1} by δ and ω^{-1} by Δ we have the 1-D realization shown in Figure 3.4



Δ : 1.01 ms delay element
 δ : 1.00 ms delay element

The output frame generator does the reverse of the ISC, displaying the picture instead of scanning.

Dimensionality of Global State

Considering a bounded frame ($M \times N$) system, it is interesting to know the dimension of the global state (or initial conditions) needed to process the $M \times N$ "future" data field. Since vertical states convey information vertically, all the vertical states along the X-axis are necessary initial conditions and their dimension is mN . Similarly, all the horizontal states along the Y-axis are necessary initial conditions (with dimension nM) since they convey information horizontally. Therefore, in the bounded frame case a total number of $mN+nM$ are needed to summarize the "past" information.

This very same idea can be used again from a computational point of view. Indeed, the number of required storage elements for recursive computations is also equal to $mN+nM$ if initial conditions are not zero. However, the initial conditions are often zero, then the size of storage required can be reduced to mN (resp. nM) by storing the updated data row by row (resp. column by column). No storage is needed for the rest of the initial conditions - nM horizontal states (resp. mN vertical states) - since they are assumed to be zero. This is consistent with the result of Read [12] derived from a direct polynomial approach.

Another interesting observation concerns the dimension of the 1-D digital filter contained in our 2-D digital filter design discussed above. Since it needs n M-unit-delays and m N-unit-delays, the corresponding 1-D state-space has also a dimension equal to $mN+nM$. Note that, despite the high dimension of the corresponding 1-D filter, its high sparsity is very encouraging for further studies.

In short, our studies on the dimensionality of 2-D global states have reached a consistent conclusion from either theoretical or practical approaches.

4. Global and Local Controllability and Observability

For reasons of space we deferred this Section to [18], part II.

5. Modal Controllability (Observability) and Minimality

In the 1-D case, the relative primeness concepts could also be used to define controllability and observability. In [16] Rosenbrock proved that

A, B was controllable if and only if $zI-A, B$ were left coprime.

C, A was observable if and only if $C, zI-A$ were right coprime.

This approach can be generalized very easily to 2-D systems and will also provide a definition of minimality.

Definition 5.1 Let $H(z, \omega) = VT^{-1}U$ where V, T, U are 2-D polynomial matrices. It is a minimal description of $H(z, \omega)$ if and only if

V, T are right coprime and T, U are left coprime.

This amounts to requiring that there is no cancellation in the 2-D transfer function $H(z, \omega)$. In [18] part I we also provide the important property that if (V, T, U) and (V_1, T_1, U_1) are two minimal descriptions of H, $|T| = |T_1|$. We also presented an algorithm to extract the greatest common right (left) divisor of two polynomial matrices, which enables us to find a minimal description of H from a nonminimal one.

Define $A(z, \omega) \triangleq \begin{pmatrix} zI_n & 0 \\ 0 & \omega I_m \end{pmatrix} - A = A_{n,m}(z, \omega)$.

Then, in the state space description case $H = CA(z, \omega)^{-1}B$ is minimal if and only if

$$[A(z, \omega), B] \text{ are left coprime} \quad (5.1)$$

and

$$[C, A(z, \omega)] \text{ are right coprime} \quad (5.2)$$

Definition 5.2 (i) A, B is modally controllable if (5.1) holds.
(ii) C, A is modally observable if (5.2) holds.

These definitions are clearly connected to minimality but the state space significance of controllability and observability disappears in this formulation. This is why we shall give now an equivalent state-space characterization of the notions of modal controllability and observability. Another consequence is that for a single input-single output system, if $H(a, \omega) = \frac{b(a, \omega)}{a(z, \omega)}$ and if b and a are coprime with $\partial_z a = n$ $\partial_\omega a = m$, then if $CA_{pq}(z, \omega)^{-1}B$ is a minimal realization of $H(z, \omega)$ we must have $|A(z, \omega)| = a(z, \omega)$ and hence $p = n$ and $q = m$.

Hence the validity of our definition of minimality of a state-space model will depend on our ability to realize a transfer function of order (n, m) with $n+m$ states. This problem was considered in Section 6 of [18], part II.

A consequence of the relative primeness criterion for 2-D polynomial matrices given in [18], part I is that C and $A(z, \omega)$ are right coprime if and only if

$$\text{rank } [A(z, \omega)] = n+m$$

for any generic point (ξ_1, ξ_2) of any irreducible algebraic curve V_1 appearing in the decomposition of V, the algebraic curve defined by $|A(z, \omega)| = 0$. It is to be noted that the rank is considered over the field $K(\xi_1, \xi_2)$. A proof of this is given in [18], part II, along with some illustrating examples.

6. Minimality of State-Space Model

It is shown in the last section and in [18], part II, that only a state-space realization with order (n, m) - i.e. the same order as the transfer function - can be both modally controllable and modally observable. Now the question is whether such a realization exists at all.

The best way to prove the existence of such realization is by construction. Note that, in the

2-D state-space model, the particular transform

$$\begin{bmatrix} \tilde{x}_h \\ \tilde{x}_v \end{bmatrix} = \begin{bmatrix} T_h & 0 \\ 0 & T_v \end{bmatrix} \begin{bmatrix} x_h \\ x_v \end{bmatrix} = T \begin{bmatrix} x_h \\ x_v \end{bmatrix} \quad (6.1)$$

enables us to change the basis of the state-space. The matrices $\{A, B, C, D\}$ are transformed to

$$\begin{aligned} \tilde{A} &= T A T^{-1} & \tilde{B} &= T B \\ \tilde{C} &= C T^{-1} & \tilde{D} &= D \end{aligned} \quad (6.2)$$

In fact, it is more convenient to work with a canonical form under the "similarity transform" defined by (6.2).

In the 1-D case, all minimal state-space model can be transformed to the controller canonical form. Similarly, almost all [18] 2-D state-space model can be transformed to the following modal controller form $\{A, B, C, \}$ (assuming $D = 0$)

$$A = \begin{bmatrix} z^{-1}e_1 - a_{n0} & : & A_{12} \\ \vdots & & \vdots \\ A_{21} & : & z^{-1}e_m - a_{0m} \end{bmatrix}, \quad B = \begin{bmatrix} e_1 \\ \vdots \\ e_m \end{bmatrix}, \quad C = \begin{bmatrix} b_{-n0} & | & b_{-0m} \end{bmatrix} \quad (6.3)$$

where z, a, b, e were defined in (3.3) and the entries of A_{12} and A_{21} are to be chosen such that

$$\det[A(z, \omega)] = a(z, \omega) \quad (6.4)$$

and

$$\det \begin{bmatrix} A(z, \omega) & B \\ -C & 0 \end{bmatrix} = b(z, \omega) \quad (6.5)$$

It is easy to check that, in (6.4), the coefficients $\{a_{i0}, 0 \leq i \leq n\}$ and $\{a_{0j}, 0 \leq j \leq m\}$ have already been matched. Similarly, in (6.5), the coefficients $\{b_{i0}, 0 \leq i \leq n\}$ and $\{b_{0j}, 0 \leq j \leq m\}$ have already been matched. Therefore, only $2nm$ coefficients $\{a_{ij}, 1 \leq i \leq n, 1 \leq j \leq m\}$ and $\{b_{ij}, 1 \leq i \leq n, 1 \leq j \leq m\}$ are to be matched. In other words, there are totally $2nm$ (nonlinear) equations to be satisfied. Coincidentally, the number of free parameters in matrices A_{12} and A_{21} is also $2nm$. Therefore it is natural to conjecture that a solution (or, more precisely, a finite number of solutions) should always exist.

Now let us examine the plausibility of this conjecture by taking a low-order example.

Example 6.1 (1,1) order case

For ease of notation, let $A_{12} = \alpha$, $A_{21} = \beta$. Also (without loss of generality) let us assume that $b_{10} \neq 0$ (otherwise, we may have to use another canonical form). Then (4) becomes

$$z\omega + a_{01} \cdot z + a_{10} \cdot \omega - d\beta = z\omega + a_{01} \cdot z + a_{10} \cdot \omega + a_{11} \quad \text{or equivalently} \quad \alpha\beta = -a_{11} \quad (6.6)$$

and (6.5) becomes

$$b_{01}z + b_{10}\omega + (a_{10}b_{01} + a_{01}b_{10} + \alpha\beta) = b_{01}z + b_{10}\omega + b_{11}$$

or

$$b_{01}\beta + b_{10}\alpha = b_{11} - a_{01}b_{10} - a_{10}b_{01} \quad (6.7)$$

Since $b_{10} \neq 0$, (6.6) and (6.7) have solutions

E. Sontag (Univ. of Florida) independently arrived at the same conjecture recently (private communication).

$$\alpha = \frac{1}{2b_{10}} (b_{10}^{-a_{01}} b_{10}^{-a_{10}b_{01}} \pm \sqrt{(b_{11} - a_{10}b_{10}^{-a_{10}b_{01}} - \dots - 4a_{11}b_{01}b_{10})})$$

$$\beta = -\frac{a_{11}}{\alpha} \quad (6.8)$$

Therefore, the existence of (1,1) order state-space model has been proved by construction. \square

Unfortunately, (6.4) and (6.5) usually give a set of $2nm$ nonlinear equations; therefore the solution may not always be in real numbers. For realization with real-gain constraints, we often need a realization order higher than $n+m$. To show that an (n,m) order real-gain realization may not exist, it is easiest to work on an example.

Example 6.2 The problem is to show that there is no (1,1) order real-gain realization for the transfer function

$$\frac{z + \omega}{z\omega - 1}$$

Solution: Let us assume

$$A = \begin{bmatrix} 0 & \alpha \\ \beta & 0 \end{bmatrix}, \quad B = \begin{bmatrix} e \\ f \end{bmatrix}, \quad C = [g, h] \quad (6.9)$$

Since $a_{11} = -1$, $\beta = \alpha^{-1}$. Then (6.4) is satisfied, and (6.5) becomes

$$fhz + eg\omega - (eh\alpha^{-1} + gfa) = z + \omega \quad (6.10)$$

or equivalently,

$$fh = 1 \quad (6.11)$$

$$eg = 1 \quad (6.12)$$

$$eh\alpha^{-1} + gfa = 0 \quad (6.13)$$

Now, (6.13) $\times hg - (6.11) \times g^2\alpha - (6.12) \times h^2\alpha^{-1}$ gives

$$g^2\alpha + h^2\alpha^{-1} = 0 \quad (6.14)$$

Since (6.14) has no real number solution, no (1,1) order real-gain realization exists, (e.g. $f = h = e = g = 1$, $\alpha = -\beta = \sqrt{-1}$). \square

In the practical aspect, real-gain realizations are much more desirable than complex realizations because the former are much easier to physically implement. Therefore our $(2m+n)$ order real-gain realization (cf. Section 3) are justified to be practical and low order realizations. Indeed, for the transfer function in Example 6.2, the minimal real-gain realization $\{A, B, C\}$ can be obtained by our realization method;

$$A = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & -1 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} 1|0|1 \end{bmatrix}, \quad C = \begin{bmatrix} 1|0|1 \end{bmatrix} \quad (6.15)$$

Special Transfer Functions

In designing digital filter, the transfer function may be intendedly chosen in a certain form for the purpose of an easier and/or better

realization. Therefore, it is worth mentioning that some special types of transfer function can be easily realized in $(n+m)$ order real-gain realizations. There are two important special types of transfer functions:

- i. with separable denominator.
- ii. with separable numerator.

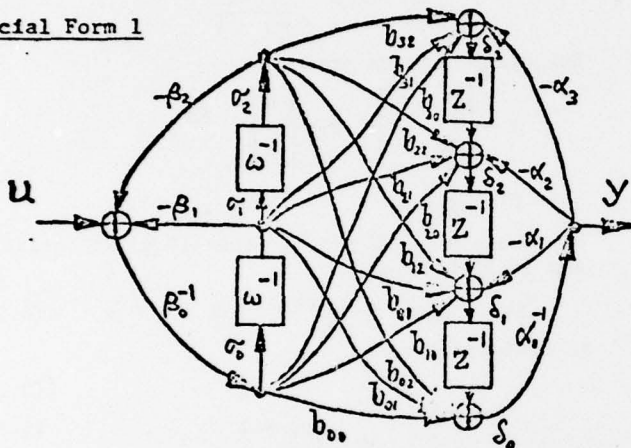
Let us first consider the separable denominator case. Assuming

$$H(z^{-1}, \omega^{-1}) = \frac{b(z^{-1}, \omega^{-1})}{a(z^{-1}, \omega^{-1})} = \frac{b(z^{-1}, \omega^{-1})}{(z^{-1}, \beta(\omega^{-1}))}, \quad \alpha_0 \neq 0 \neq \beta_0 \quad (6.16)$$

$$= \frac{\sum_{i=0}^n \sum_{j=0}^m b_{ij} z^{-i} \omega^{-j}}{(\alpha_0 + \alpha_1 z^{-1} + \dots + \alpha_n z^{-n})(\beta_0 + \beta_1 z^{-1} + \dots + \beta_m z^{-m})}$$

then its circuit realization is shown in Figure 6.1

Special Form 1



Controller-Observer-Type Form

Secondly, let us consider the separable numerator case, which is to say a system with transfer function.

$$\tilde{H}(z^{-1}, \omega^{-1}) = [H(z^{-1}, \omega^{-1})]^{-1} = \frac{\alpha(z^{-1}) \beta(\omega^{-1})}{\sum_{i=0}^n \sum_{j=0}^m b_{ij} z^{-i} \omega^{-j}} \quad (6.17)$$

At first sight, it seems quite difficult. However, in actuality, the realization can be readily obtained by using the inversion rule by Kung [17]. More precisely, to realize the inverse system of Figure 6.1, we first note that the path "input -- α_0 -- δ_0 -- input" is a "feed through" path (i.e. a path connecting input and output with only constant gains). The second step is to invert all the gains and reverse all the arrows on the path (in our case, replace b_{00} by b_{00}^{-1}). Lastly, change signs of the gains of the branches which are entering this path. These steps complete the realization of $H^{-1}(z^{-1}, \omega^{-1})$, in [18], part II the implementation is given.

Remark: In many design problems the constraints on numerator are much weaker than on denominator, hence this second form seems to have higher potential in practical applications.

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Appendix: 2-D Levinson Algorithms: The following set of results were motivated by the problem of determining stability of 2-D recursive filters.

In the 1-D case the connections between stability, orthogonal polynomials and the Levinson recursions are by now well known.

In the 2-D case, Shank conjectured that the least-squares inverse, say $b(z, \omega)$, of an unstable 2-D polynomial $a(z, \omega)$ (of degree n in z , m in ω) is stable, i.e.,

$$\sum_{i,j=0}^{n,m} b_{ij} z^i \omega^j = b(z, \omega) \simeq 1/a(z, \omega), \quad (A.1)$$

where $b(z, \omega)$ minimizes

$$\|1 - a(z, \omega)b(z, \omega)\|^2 \quad (A.2)$$

or

$$\|e_{00} - a b\|^2$$

with

$$e'_{00} \triangleq [1, 0, \dots, 0], [1 \times (2n+1)(m+1)]$$

$$b' = [b_{00}, \dots, b_{0m}, b_{10}, \dots, b_{nm}]$$

and the Toeplitz block Toeplitz matrix a containing the coefficients of $a(z, \omega)$ such that $[a b]$ is the vector of the coefficients of the product polynomial $[a(z, \omega)b(z, \omega)]$. Then b is given by the solution of

$$a b = [a' a] b = e_{00} a_{00} \quad (A.3)$$

By applying the Levinson (LRN) recursions [19], for block matrices developed by Robinson and Wiggins [20], to (A.3) b can be obtained from the first column of the block solution $a B = [I_m, 0, \dots, 0]'$. Or with

$$\begin{aligned} \underline{\omega} &\triangleq [1, \omega, \dots, \omega^m]' \\ b(z, \omega) &= \sum_{i=0}^m b_i(z) \omega^i = \underline{\omega}' \underline{b}(z) \quad (A.4) \\ &= [\underline{\omega}', \underline{\omega}' z, \dots, \underline{\omega}' z^n] \underline{B} \underline{e}_1 = \underline{\omega}' B(z) \underline{e}_1. \end{aligned}$$

Using the property of the LRN recursions that $|B(z)|$ has its roots inside the unit circle, the g.c.d. of $\{b_i(z)\}$, which divides $|B(z)|$ contains a subset of these roots.

We can therefore conclude that the nonprimitive factors of $b(z, \omega)$ --the contents in z and ω --are indeed stable.

However Genin and Kamp [21] proved that $b(z, \omega)$, therefore the primitive factors, are in general not stable for $n, m > 1$.

A 2-D Levinson Algorithm: Genin and Kamp developed a 2-D generalization of the orthogonal polynomials on the unit circle. We give here an equivalent recursion in the time-domain using a stochastic framework (see, e.g., [19]).

We consider a finite window of a scalar 2-D stationary stochastic process $\{y_{ij}, i \in [0, n], j \in [0, m]\}$ with zero mean and covariance

$$E(y_{ij} y_{hk}) = r_{i-h, j-k}.$$

A. Daggroer (MIT) independently also developed such recursions (private communication).

Define

$$y_j^{(n,m)} \triangleq [y_{j,m}, \dots, y_{j,0}]'$$

as the j th column of the data array, $0 \leq j \leq n$ and

$$y^{(n,m)} \triangleq [y_n^{(n,m)}, \dots, y_0^{(n,m)}]'$$

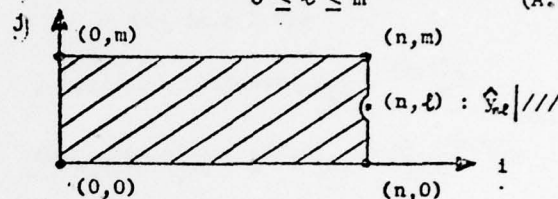
the data array scanned column by column. Now the covariance of $y^{(n,m)}$ is given by

$$E(y^{(n,m)} y^{(n,m)}) = \mathcal{Q}^{(n,m)} \quad (A.5)$$

where $\mathcal{Q}^{(n,m)}$ is a $(n+1)$ by $(n+1)$ block Toeplitz matrix with Toeplitz block entries, $R_{j-i} = [\mathcal{Q}^{(n,m)}]_{i,j}$ of size $(m+1)$ by $(m+1)$ and $R_{-k} = R_k'$, $\{R_k\}_{ij} = r_{k,j-i}$. Now, let

$$\hat{y}(n, \ell; n, m) = E(y(n, i) | y(r, s) :$$

$$(0, 0) \leq (r, s) \leq (n, m), (r, s) \neq (n, \ell), 0 \leq \ell \leq m \quad (A.6)$$



$$\hat{y}(n, \ell; n, m) = -y^{(n,m)} \mathcal{Q}_{(n, \ell)}^{(n,m)}, \quad (A.7)$$

where

$$\mathcal{Q}_{(n, \ell)}^{(n,m)} \triangleq [\mathcal{Q}_0^{(n,m)}(n, \ell), \dots, \mathcal{Q}_n^{(n,m)}(n, \ell)]'$$

and

$$\mathcal{Q}_0^{(n,m)}(n, \ell) = [x, \dots, x, 0, x, \dots, x] \quad \text{\scriptsize $\ell m - \ell + 1$ position}$$

So that if

$$\tilde{y}_c^{(n,m)} = [\tilde{y}(n, m; n, m), \dots, \tilde{y}(n, \ell; n, m), \dots, \tilde{y}(r, s; n, m)]$$

$$\tilde{y}(n, \ell; n, m) = y(n, \ell; n, m) - \hat{y}(n, \ell; n, m)$$

Then

$$\tilde{y}_c^{(n,m)} = -y^{(n,m)} \mathcal{Q}_c^{(n,m)} \quad (A.8)$$

where

$$\mathcal{Q}_c^{(n,m)} = \underline{e}_1 \otimes I_{m+1} + [\mathcal{Q}_{(n,m)}^{(n,m)} \dots \mathcal{Q}_{(n,\ell)}^{(n,m)} \dots \mathcal{Q}_{(n,0)}^{(n,m)}] \quad (A.9)$$

Note that diagonal entries of top block of $\mathcal{Q}_c^{(n,m)}$ equal unity, and \otimes denotes Kronecker product. Similarly we can define

$$\hat{y}(k, m; n, m) = E(y(k, m) | y(r, s) :$$

$$(0, 0) \leq (r, s) \leq (n, m), (r, s) \neq (k, m) = \quad (A.6')$$

$$= -y^{(n,m)} \mathcal{Q}_{(k, m)}^{(n,m)}, 0 \leq k \leq n \quad (A.7')$$

also

$$\tilde{y}_r^{(n,m)} = [\tilde{y}(n, m; n, m), \dots, \tilde{y}(k, m; n, m), \dots, \tilde{y}(0, m; n, m)],$$

$$\tilde{y}(k, m; n, m) = y(k, m; n, m) - \hat{y}(k, m; n, m)$$

and

$$\tilde{y}_r^{(n,m)} = -y^{(n,m)} a_r^{(n,m)} \quad (A.8')$$

where

$$Q_r^{(n,m)} = I_{n+1} \otimes \underline{e}_1 + [a_{(n,m)}^{(n,m)} \dots a_{(k,m)}^{(n,m)} \dots a_{(0,m)}^{(n,m)}] \quad (A.9')$$

Note that diagonal entries of block composed of i th row of i th block entry of $Q_r^{(n,m)}$ equal unity. Also, the first columns of $Q_c^{(n,m)}$ and of $Q_r^{(n,m)}$ are the same. Then by (A.5), (A.8) and (A.8')

$$E[y^{(n,m)} [\tilde{y}_r^{(n,m)}, \tilde{y}_c^{(n,m)}]] = Q_r^{(n,m)} [a_r^{(n,m)}, a_c^{(n,m)}]$$

and also by (A.6) and (A.6')

$$E[y^{(n,m)} [\tilde{y}_r^{(n,m)}, \tilde{y}_c^{(n,m)}]] = [\mathcal{E}_r^{(n,m)}, \mathcal{E}_c^{(n,m)}],$$

$$\mathcal{E}_r^{(n,m)} \triangleq \text{Diag}[\epsilon(n-1, m; n, m)] \otimes \underline{e}_1, [(n+1) \times (n+1)] \\ \epsilon(k, m; n, m) \geq 0 \\ \epsilon(n, l; n, m) \geq 0.$$

$$\mathcal{E}_c^{(n,m)} \triangleq \underline{e}_1 \otimes \text{Diag}[\epsilon(n, m-1; n, m)],$$

hence

$$R^{(n,m)} [a_r^{(n,m)}, a_c^{(n,m)}] = [\mathcal{E}_r^{(n,m)}, \mathcal{E}_c^{(n,m)}] \quad (A.10)$$

are the 2-D Levinson equations, therefore we have $n+m+1$ auxiliary solutions. Also, the first column of $Q_c^{(n,m)}$ (or of $Q_r^{(n,m)}$ since they are the same) corresponds to the 2-D causal estimate of $y(n, m)$ given $y(i, j) : (0, 0) \leq (i, j) < (n, m)$, i.e., it is the $++$ predictor of $y(n, m)$ (one quadrant-predictor). The last column of $Q_r^{(n,m)}$ gives the $-+$ predictor and the last column of $Q_c^{(n,m)}$ gives the $+-$ predictor.

The Levinson Recursions: First define

$$[J_m]_{ij} \triangleq \begin{cases} 1 & \text{if } i=m-j \\ 0 & \text{else} \end{cases}, \quad 0 \leq i, j \leq m$$

Now, observe that the following reorderings hold:

$$(J_n \otimes J_m) R^{(n,m)} (J_n \otimes J_m) = R^{(n,m)}, \quad (A.11)$$

hence

$$(J_n \otimes J_m) R^{(n,m)} (J_n \otimes J_m) \underbrace{(J_n \otimes J_m)}_{I_{(n+1)(m+1)}} [a_r^{(n,m)}, a_c^{(n,m)}] \\ = (J_n \otimes J_m) [\mathcal{E}_r^{(n,m)}, \mathcal{E}_c^{(n,m)}],$$

so that

$$R^{(n,m)} (J_n \otimes J_m) [a_r^{(n,m)}, a_c^{(n,m)}] = [\mathcal{E}_r^{(n,m)}, \mathcal{E}_c^{(n,m)}]$$

and we multiply

$$\begin{pmatrix} J_n & 0 \\ 0 & J_m \end{pmatrix}$$

on the right and denote

$$a_r^{*(n,m)} = (J_n \otimes J_m) a_r^{(n,m)} J_n,$$

$$a_c^{*(n,m)} = (J_n \otimes J_m) a_c^{(n,m)} J_m$$

and

$$\mathcal{E}_r^{*(n,m)} = (J_n \otimes J_m) \mathcal{E}_r^{(n,m)} J_n,$$

$$\mathcal{E}_c^{*(n,m)} = (J_n \otimes J_m) \mathcal{E}_c^{(n,m)} J_m.$$

Then

$$R^{(n,m)} [a_r^{*(n,m)}, a_c^{*(n,m)}] = [\mathcal{E}_r^{*(n,m)}, \mathcal{E}_c^{*(n,m)}] \quad (A.12)$$

and

$$\begin{aligned} \mathcal{E}_r^{*(n,m)} &= J_n \text{Diag}[\epsilon(n-1, m; n, m)] J_n \otimes \underline{e}_{m+1} = \\ &= \mathcal{E}_r^{*(n,m)} \otimes \underline{e}_{m+1} \end{aligned}$$

(we have used $(A \otimes B)(C \otimes D) = AC \otimes BD$). Similarly

$$\begin{aligned} \mathcal{E}_c^{*(n,m)} &= \underline{e}_{m+1} \otimes J_m \text{Diag}[\epsilon(n, m-1; n, m)] J_m = \\ &= \underline{e}_{m+1} \otimes \mathcal{E}_c^{*(n,m)} \end{aligned}$$

Now define

$$\mathcal{Q}_c^{(n,m)} \triangleq [R_{-n-1}, R_{-r}, \dots, R_{-1}] a_c^{(n,m)} \quad (A.13)$$

$$\beta_r^{(n,m)} \triangleq [R_1, \dots, R_n, R_{n+1}] a_r^{(n,m)}. \quad (A.14)$$

Now, the 2-D Levinson recursions can be described as follows. Increase in $n: n \rightarrow n+1, m \rightarrow m$, see Figure 41.

$$R^{(n+1,m)} \begin{bmatrix} a_c^{(n,m)} \\ 0_{m+1} \end{bmatrix} = \begin{bmatrix} \mathcal{E}_c^{(n,m)} \\ \mathcal{Q}_c^{(n,m)} \end{bmatrix} \begin{bmatrix} a_c^{*(n,m)} \\ \mathcal{E}_c^{*(n,m)} \end{bmatrix}$$

$$(\mathcal{Q}_c^{*(n,m)} = J_m \mathcal{Q}_c^{(n,m)} J_m).$$

Now, let

$$\begin{aligned} \bar{a}_c^{(n+1,m)} &= \begin{pmatrix} a_c^{(n,m)} \\ 0 \end{pmatrix} - \\ &- \begin{pmatrix} 0 \\ a_c^{*(n,m)} \end{pmatrix} E_c^{*(n,m)-1} \mathcal{Q}_c^{(n,m)-1} \Lambda_c^{(n,m)-1} \end{aligned} \quad (A.15)$$

where

$$\Lambda_c^{(n,m)} = E_c^{(n,m)} - \mathcal{Q}_c^{*(n,m)} E_c^{*(n,m)-1} \mathcal{Q}_c^{(n,m)}.$$

Then

$$R^{(n+1,m)} \bar{a}_c^{(n+1,m)} = \underline{e}_1 \otimes I_{m+1} \quad (A.16)$$

and the diagonal of the top block of $\bar{a}_c^{(n+1,m)}$ equals $\text{Diag}[\epsilon(n+1, m-1; n+1, m)]^{-1}$ so that $a_c^{(n+1,m)}$ is just obtained by re-normalizing the columns of

$$\bar{Q}_c^{(n+1,m)}$$

Note 1: $\epsilon^{-1}(n+1,k;n+1,m) \neq 0$ otherwise deleting the k th column and k th row of $\bar{Q}^{(n+1,m)}$, we could find ϵ_k : $\bar{Q}_k^{(n+1,m)} \epsilon_k = 0$. But $\bar{Q}_k^{(n+1,m)}$ is a covariance and this would mean that the estimation problem is singular.

Note 2: Similarly $\epsilon(n+1,k;n+1,m) \neq 0$ otherwise there would be ϵ : $\bar{Q}^{(n+1,m)} \epsilon = 0$. Also

$$\bar{Q}^{(r+1,m)} \begin{bmatrix} 0_{n+1}^{m+1} \\ \bar{Q}_c^{(n+1,m)} \\ Q^{(r+1,m)} \end{bmatrix} = \begin{bmatrix} \bar{P}_r^{(n,m)} & I_{m+1} \\ \epsilon_r^{(n,m)} & 0 \end{bmatrix}$$

and let

$$\check{Q}_r^{(n+1,m)} = \begin{bmatrix} 0_{n+1}^{m+1} \\ Q_r^{(n,m)} \end{bmatrix} - \bar{Q}_c^{(n+1,m)} \bar{P}_r^{(n,m)} \quad (A.17)$$

then

$$Q_r^{(n+1,m)} = \begin{bmatrix} Q_{c1}^{(r+1,m)} & \check{Q}_r^{(n+1,m)} \end{bmatrix}, \quad (n+2 \text{ columns}) \quad (A.18)$$

where $Q_{c1}^{(n+1,m)}$ is the first column of $Q_c^{(n+1,m)}$.

To obtain the recursions for an increase in m , we just have to reorder $\bar{Q}^{(n,m)}$ in blocks of size $(n+1) \times (n+1)$ then the roles of m and n are exchanged as well as Q_c and Q_r and we can use the same recursion as the one just described.

This version of the recursion enables us to increase n and m separately, instead of the scheme proposed by Genin and Kamp where $m = n$.

Note 3: The inversions required by these recursions have additional structure, i.e., the matrices are typically non-Toeplitz, but sums of products of Toeplitz matrices. One can take advantage of such a structure by using generalized Levinson recursions [22] to find a representation of the inverse of such matrices also in terms of sums of products of Toeplitz matrices. Expressions with Toeplitz matrices, since they are related to convolutions, can be evaluated using Fast Fourier Transforms (FFT's).

$$\begin{bmatrix} R_0 & R_1 & \dots & R_n & R_{n+1} \\ R_1' & R_0 & \dots & R_{n-1} & R_n \\ \dots & \dots & \dots & \dots & \dots \\ R_n' & R_{n-1}' & \dots & R_0 & R_1 \\ R_{n+1}' & R_n' & \dots & R_1' & R_0 \end{bmatrix} \cdot \begin{bmatrix} 0 & I_{n+1} \\ I_{n+1} & X \\ \dots & \dots \\ X & X \\ X & X \\ X & 0 \end{bmatrix} = \begin{bmatrix} \bar{P}_r^{(n,m)} & 0 \\ \Delta_r & 0 \\ \dots & \dots \\ 0 & 0 \\ 0 & Q_c^{(n,m)} \end{bmatrix} \quad \left. \begin{array}{l} \} \epsilon_c^{(n,m)} \\ \} \{\Delta_r\} : \epsilon_r^{(n,m)} \end{array} \right\}$$

FIGURE A1.

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20. ABSTRACT (Continue on reverse side if necessary and identify by block number) A short comparison between the different state-space models is presented. We discuss proper definitions of state, controllability and observability and their relation to minimality of 2-D systems. We also present new circuit realizations and 2-D digital filter hardware implementations of 2-D transfer functions, as well as a 2-D generalization of Levinson's algorithms.		